The thermodynamics of lipid ion channel formation in the absence and presence of anesthetics. BLM experiments and simulations†

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It is known that lipid membranes become permeable in their melting regime. In conductance measurements on black lipid membranes one finds that conduction takes place *via* quantized events closely resembling those reported for protein ion channels. Here, we present data of ion currents through black lipid membranes in the presence and absence of the anesthetics octanol and ethanol, and compare them to a statistical thermodynamics model using parameters that are obtained from experimental calorimetric data. The conductance steps in pure lipid membrane suggest aqueous pores with the size of approximately one lipid cross-section. We model the permeability by assuming empty lattice sites of the size of one lipid. We find that pore formation in the melting transition regime is facilitated by the increase of the lateral compressibility that expresses itself in the area fluctuations, a process that is related to critical opalescence in two dimensions. Anesthetics alter the permeability by affecting the thermodynamic state of the membrane and by shifting the heat capacity profiles.

1 Introduction

In biology, lipid membranes are typically considered as insulators, which is an important assumption for models related to nerve pulse propagation (e.g., the Hodgkin-Huxley model¹) and the conduction of ions through proteins. It is known, however, that lipid membranes become permeable in the proximity of membrane melting transitions,2-4 which creates significant conceptual problems for such models because biological membranes under physiological conditions are in fact often close to such transitions. We have shown in two recent papers that the membrane permeability changes continuously in the transition with a behavior closely resembling the heat capacity anomaly.^{6,7} This is both true for permeation of fluorescence dyes⁶ and for potassium and sodium chloride^{6,7} through black lipid membranes of lipid mixtures in their transition regime. The increased permeability of membranes in transitions has been attributed to the large fluctuations in membrane area generally related to transitions, and to the appearance of domain boundaries.^{2,8} The area fluctuations are proportional to the lateral compressibility9 that also reaches a maximum in the transition regime. Thus, the work necessary to create conducting membrane defects under these conditions is small.3

Interestingly, it has been found in microscopic measurements using black lipid membranes or using a patch-clamp that the conductance of ions through pure lipid membranes occurs in quantized steps, suggesting pores of well-defined size.^{6,10–15} These events display a similar conductance and typical opening and closing time scales as the conduction events reported for protein ion channels.¹⁶ In a recent paper we found conductance steps

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suggesting aqueous pores of 0.7 nm, very similar to the size proposed for protein ion channels.⁶ Pores of this size have also been proposed on the basis of molecular dynamics simulations.¹⁷

The obvious question arises how the lipid conduction events can be distinguished from those of proteins. In particular, while lipid membranes can be investigated in the absence of proteins, channel proteins cannot be investigated in the absence of lipid membranes into which they are embedded. A patch pipette has a diameter of about 1 µm while the diameter of a typical membrane protein is of the order of 5 nm. The patch tip area is therefore about 40 000 times larger than the protein, which renders it technically impossible to measure currents through proteins without simultaneously measuring the events in the lipid membrane, too.

Taking the very convincing evidence of a direct coupling between heat capacity and permeability we concluded that everything shifting the heat capacity profile of membrane melting will also alter the membrane permeability. This includes changes in temperature, pressure and pH, but also the presence of anesthetics. The latter molecules have been shown to lower membrane transitions according a freezing point depression law, completely independent of the nature of the anesthetics drug. When measuring in the transition regime, anesthetics therefore lower the permeability, and quantized current events are 'blocked'. F

In this paper we show experimental data on quantized currents through black lipid membranes (BLMs) and the effect of anesthetics on the conduction. In comparison to our two earlier recent publications^{6,7} we focus here on investigating the origin of the current fluctuations from lateral area fluctuations that are strongly enhanced in transition regimes. We rationalize our findings with a statistical thermodynamics model that takes the area fluctuations of the membranes into account. Statistical thermodynamics models (e.g., 2 state Ising models or 10 state Pink models) in the past have been successfully used to describe cooperative melting events in membranes.^{19–25} In transitions, the cooperative

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fluctuations lead to domain formation on scales much larger than the individual lipid. On these scales, the lack of molecular detail is not important. For this reason, one cannot deduce molecular pore geometries. In this publication we allow for the formation of pores facilitated by the fluctuations in area without specifying a detailed pore geometry. We demonstrate that this models results in channel statistics similar to the BLM experiment and that this approach explains the effect of anesthetics.

2 Methods

2.1 Materials

Chemicals. Decane and 1-octanol were purchased from Fluka Chemie AG (Deisenhofen, Germany), N-hexadecane, chloroform and ethanol were obtained from Merck (Hohenbrunn, Germany), n-pentane was provided by BDH (Poole, UK) and potassium chloride from J.T. Baker Analyzed (Deventer, Holland). 1,2-dipalmitoyl-sn-glycero-3-phosphocholine (DPPC) and 1,2-dioleoyl-sn-glycero-3-phosphocholine (DOPC) were purchased from Avanti Polar Lipids (Birmingham, AL) and used without further purification. For all experiments, MilliQ water (18.1 M Ω) was used.

Calorimetry. DSC experiments were performed on a VP-DSC Calorimeter (MicroCal, Nort-hampton, MA, USA) with a scan rate of 5 °C/h. The lipid samples were prepared by pre-dissolving in chloroform, which was first dried under a nitrogen stream and then kept under high vacuum overnight. The dried lipid mixtures were dispersed in MilliQ water to a final concentration of 20 mM. The buffer used was the same as for the BLM experiments. Before filling the calorimeter, the solutions were degassed for 10 min.

Black lipid membranes. Planar bilayers were formed over a round aperture of ~80 μm radius in a Teflon film of 25 μm thickness dividing two compartments of a Teflon chamber embedded in a brass block that could be heated by a circulating water bath. The aperture in the Teflon film was prepainted with 5% hexadecane in pentane. The BLMs were painted with DOPC:DPPC 2:1 lipid solutions in decane/chloroform/methanol 7:2:1 and formed following the method described by Montal and Mueller.²⁶ The two compartments of the Teflon block were filled with unbuffered 150 mM KCl (pH 6.5). Lipid solution (25 mg/ ml) was spread on the buffer surface in each compartment (approx. 3 µl on each side). Ag/AgCl electrodes were placed into both compartments of the chamber. After 15-30 min to allow for the evaporation of the solvent, the water level of the compartments was lowered and raised several times until a bilayer was formed over the hole. The formation of BLMs was controlled visually and by capacitance measurements (with a triangular 100 mV voltage input pulse). For the experiments with octanol, 15% v/v octanol in methanol solution was prepared and added symmetrically on both sides of the Teflon chamber. To calculate the concentration of octanol in the membrane we assumed an octanol partition coefficient between lipid membrane and water of 200.27 In the experiments with ethanol we used 60 µl and 120 µl of 99% ethanol for a buffer volume of the two compartments of 2 ml. Assuming a ethanol partition coefficient of 0.48 between

the membranes and water²⁸ this corresponds to 9.5 mol% and 19 mol% ethanol in the membrane.

Conductance measurements were performed on an Axopatch 200B amplifier in voltage clamp mode connected to a DigiData 1200 digitizer (both Molecular Devices, Sunnyvale, CA, USA). Current traces were filtered with 1 kHz low-pass Bessel filter and recorded with Clampex 9.2 software (Axon Instruments) on the hard drive of the computer using an AD converter with a time resolution of 0.1 ms. The data was further analyzed with Clampfit 9.2 and low-pass filtered with Bessel (8-pole) filter at a cut-off frequency of 300 Hz. Temperature was controlled by a HAAKE DC30 K20 (Waltham, USA) waterbath and a thermocouple (WSE, Thermocoax).

2.2 Simulating pore formation

Here we introduce a statistical mechanical model for pore formation close to lipid melting transitions in a system with one species of lipid and a second species of molecule (here anesthetics), based on previous work by Ivanova *et al.*²⁵ We further base our simulations on a theoretical paper by Nagle and Scott³ that relates pore formation to the lateral compressibility of the membrane. The aim of this work was to rationalize the findings from a paper by Papahadjopoulos *et al.*² that has first shown experimentally that membranes are most permeable in their transition. According to Nagle and Scott, the work necessary to compress a membrane (*i.e.*, the Helmholtz free energy change) by an area *a* is a quadratic function of the area that has the form

$$\Delta W(a) = \frac{1}{2} K_T^A \left(\frac{a}{A_0}\right)^2 \cdot A_0 \tag{1}$$

where A_0 is the total area of the membrane, a is the size of a pore defect and K_T^4 is the isothermal area compression modulus (e.g., ref. 3). The compressibility of a membrane is the inverse of the compression modulus given by

$$\kappa_T^A = -\frac{1}{A_0} \left(\frac{\partial A}{\partial \Pi} \right)_T \equiv \frac{1}{K_T^A}$$
(2)

Nagle and Scott³ assumed that the permeability can be expressed as a series expansion with respect to the area fluctuations that are also proportional to the lateral compressibility. The resulting permeability obtained by Nagle and Scott is

$$P = a_0 + a_2 \kappa_T^A + \text{higher order terms}$$
 (3)

where a_0 and a_2 are constants to be determined from experiment. We have shown previously that close to transitions the lateral compressibility is proportional to the heat capacity:

$$\kappa_T^A = \kappa_{T,0}^A + \frac{\gamma_A^2 T}{A_0} \Delta c_p \tag{4}$$

Inserting expression 4 into eqn 3 yields

$$P = c_0 + c_2 \Delta c_p \tag{5}$$

where c_0 and c_2 are also constants to be determined from the experiment. This is a simple expression directly coupled to the

calorimetric experiment. More details can be found in ref. 6. As pointed out in ref. 6, this approach resembles that employed by Einstein to rationalize critical opalescence in lipid mixtures close to a critical point.²⁹ The Nagle–Scott model³ combined with eqn 4 as given by eqn 5 suggests a proportional relation of lateral compressibility and pore formation probability. The argument implied in this notion is that one needs to perform work against the lipid matrix to create a hole by compressing the rest of the matrix. This also implies, that the formation of the pore must keep the overall area of the membrane constant, i.e., the area of the membrane does not increase upon the introduction of a pore. Since the compressibility is at a maximum in the transition regime, one expects the highest likelihood of pore formation. Would the formation of a hole in the lipid matrix just increase the overall area one would not find a maximum in conductance in the transition because the membrane is not compressed. One would rather find a stepwise change in conductance from gel to fluid membranes. In a recent publication⁶ we have shown experimentally by fluorescence correlation spectroscopy that the permeability of a lipid membrane for a fluorescence dye in fact follows the predictions of the Nagle-Scott model in combination with the proportional relation of compressibility and heat capacity (eqn 3 and 5) within experimental accuracy. Thus, it has been experimentally shown that the pore formation process maintains the membrane area. Therefore, we designed a simulation in which the pore formation process was chosen such that the overall area was maintained.

Fig. 1a and 1b show two different possible pore geometries. They correspond to two pore intermediates found in MD simulations.¹⁷ Fig. 1b also corresponds to the geometry assumed in elasticity theory.30 In both papers, the pore size was found to be of the order 0.7-1 nm. Fig. 1a corresponds just to a vacancy of diameter of one lipid in which lipid chains are exposed to water with an associated free energy change due to the hydrophobic contact, while the second geometry (Fig. 1b) requires a rearrangement of lipids such that no hydrophobic contact with the

water exists. One has, however, a free energy of curvature for the lipids at the interface of the pore. Thus, both geometries consist of a pore with diameter of 0.7 nm and an interfacial free energy. In this work we employ a lattice Ising model where pores are represented by a free lattice site and the pore free energy by the interaction of the pore with the lipids (cf., Fig. 1c). This is a reasonable model for the both pore geometries given in Fig. 1a and b. One should keep in mind that the Ising model does not contain a detailed molecular model for the pore. It contains, however, the measured pore area, and the interfacial free energy of the pore with the lipids. Thus, representing a pore by a vacancy in a lattice does not specify a molecular geometry.

In the simulation, each lipid could be in either gel or fluid state. This model formally corresponds to an Ising model in a field. In the present work we additionally allowed for the formation of pores in the membrane by moving lipids from randomly chosen lattice sites to the edge of the computer matrix. For a given temperature we evaluate the mean number of gel and fluid lipids and the fluctuations in their number, and we determine the number of spontaneously formed pores. A numerical evaluation by means of Monte Carlo simulations was used. To do this one first has to define the various Monte Carlo steps. Whether a given step is accepted or not, depends on difference in the free energy of the old and the new system configuration as given by eqn (1). At its most basic (i.e. a one-component system without anesthetics and no pore formation) only one type of step is needed, namely one that allows the lipids to change their state (a "melting step"). As we are using the two-state Ising model as a basis, each lipid is assumed to be in either an ordered (gel) or disordered (fluid) state. 22,25 For the more advanced models (multiple lipid species, anesthetics, pore formation, etc.) it is also necessary to include the possibility for any two particles to swap positions (i.e., a "diffusion step"),31,32 as well as to include a Monte Carlo step which allows for the creation/sealing of a pore. The basic idea behind the pore forming step is the observation that a lipid changes area by approximately 25% when it goes from the fluid

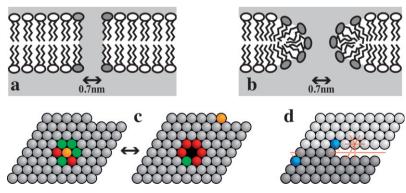


Fig. 1 a. The pore modeled by a hole in the membrane of about a lipid diameter with the chains exposed to water. The lipids with altered free energy are displayed with gray headgroups. b. The pore modeled with lipid arrangement. The head groups point towards the aqueous pore, but the monolayers display curvature. The lipids with altered free energy are displayed with gray headgroups. c. The pore formation step in the lattice Monte Carlo simulation. Green represents fluid state lipids, while the gels are red. The yellow particle can be either a lipid or an anesthetic molecule. In this Monte Carlo step three lipids simultaneously change state while the yellow particle is moved to or from the edge of the lattice. This means that the pore formation step increases the lattice size by one. During this step the overall area of the matrix stays constant (see text). d. The dark and light grey regions (bottom and top half) indicate the boundaries of repeating lattices. The definition of the nearest neighbors of the particle moved from the site of the pore to the edge. The red lines indicate the definition of nearest neighbors at the edge. Due to the odd shape of the lattice the blue lattice sites display a different definition of nearest neighborhood (indicated by the dotted red line). Still all lattice sites have 6 nearest neighbors.

to the gel state. So if three neighboring fluid lipids simultaneously change states, one can create a pore which has an area that is equal to a gel state lipid. Thus this step both conserves the number of lipids and also the area as $A_{before} = 3A_{fluid} = A_{after} = 3A_{gel} + A_{pore}$. In this kind of system a pore has automatically the cross-section of one lipid, which is what is observed in experiments⁶ and molecular dynamics simulations.¹⁷

First pore formation is described. This step can be applied if a matrix site (that is not already a pore) is randomly picked: first the surroundings (nearest neighbors) is checked for lipids in the fluid state. If there are three or more, a pore is temporarily created by changing the three fluids into gels, and moving the "central particle" to the "end" of the lattice, increasing the number of lattice sites by one (see Fig. 1, left and center). In order to obey detailed balance, there must also be a possibility to seal a pore. This can happen when a pore is picked, and it has three or more neighbors which are gel state lipids. If so, the pore is temporarily sealed by changing the three gels to fluids, and moving the "end particle" into the pore, thus reducing the number of lattice sites by one (see Fig. 1). Finally, the proposed change is accepted or rejected using the Glauber algorithm.³³ We used periodic boundary conditions using a special definition of the nearest neighbor lipid environment on the edge (see Fig. 1, right).

Model parameters. There are three parameters which need to be determined if one wants to describe the melting transition of the most basic one-component two-state lipid system, namely the melting enthalpy, ΔH , the melting entropy, ΔS , and the cooperativity parameter, ω_{fg} . The latter parameter describes the interaction between a gel and a fluid lipid and determines the half width of the melting transition. For the model additionally containing anesthetics and pores, another five interaction parameters are needed, namely ω_{af} (anesthetic/fluid), ω_{ag} (anesthetic/gel), ω_{ap} (anesthetic/pore), ω_{pg} (pore/gel), and ω_{pf} (pore/fluid).²⁴

With these parameters given the changes in free energy can be calculated as

$$\Delta G = \Delta H - T\Delta S + \Delta N_{fg}\omega_{fg} + \Delta N_{af}\omega_{af} + \Delta N_{ag}\omega_{ag} + \Delta N_{pf}\omega_{pf} + \Delta N_{pg}\omega_{pg} + \Delta N_{ap}\omega_{ap}$$
 (6)

Fortunately, the first three parameters can be obtained from calorimetric experiments, ²⁵ while the remaining three parameters are chosen such that pore formation likelihoods from permeation measurements are well described. Specifically, one can obtain the melting enthalpy and entropy of the lipids from the heat capacity profile of the one-component lipid system: the enthalpy change of the transition is obtained by integrating over the excess heat capacity in the relevant temperature interval. The entropy change can then easily be determined via $\Delta S = \Delta H/T_m$, as the transition midpoint, T_m , can be read off the heat capacity profile directly. The cooperativity parameter, ω_{fg} , relates directly to the width of the transition peak, and can consequently be determined by fitting the simulated heat capacity profiles to the experimentally measured one.

The five remaining interaction parameters, ω_{ij} , can be obtained indirectly from various other experiments. We have shown before that most anesthetics generate a freezing point depression

that is consistent with ideal solubility in the fluid phase and no solubility in the gel phase. ¹⁸ Therefore we chose $\omega_{af}=0$, and a high value for the anesthetic–gel interaction with $\omega_{ag}=\omega_{fg}$. ²⁵ Determination of the remaining three parameters (ω_{ap} , ω_{pg} and ω_{pf}) can, in principle, be done by comparing the simulated permeability (basically the average number of pores) to the measurements.

Even before doing this, it is possible to predict a number of relations, if we assume that hydrophobic matching is a major determinant of the nearest neighbor interactions. Firstly, if the added anaesthetic is strongly hydrophobic, it seems reasonable to assume that $\omega_{ap} \gg \omega_{fg}$, as the cooperativity parameter, ω_{fg} , is largely determined by the hydrophobic mismatch between a fluid and gel state lipid. Similarly, this should mean that $\omega_{pg} > \omega_{pf}$ for two reasons: (1) the lipid in the gel state would have a larger part of the hydrophobic chains exposed to the water, and (2) the fluid phase is more loosely packed, meaning that defects and spaces between the lipids (*i.e.* pores) would be more likely to appear than in the tightly packed gel phase.

Based on the considerations above, the parameter values used in the simulations were as follows: $\Delta H = 36400$ J/mol, $\Delta S = 115.87$ J/(mol K), $\omega_{fg} = 1326$ J/mol, $\omega_{pf} = 3700$ J/mol, $\omega_{pg} = 6000$ J/mol, $\omega_{pa} = 13$ 260 J/mol, $\omega_{af} = 0.0$ J/mol, and $\omega_{ag} = 1326$ J/mol. The values for ΔH , ΔS and ω_{fg} are identical to those used in ref. 25. As seen in the results section, the average number of pores formed in the BLM experiments is in the range 1-2 in an area of 50 μm^2 corresponding to 10^8 lipid molecules. In our simulations we used a matrix with 10 000 lipids that is 10 000 times smaller than the experimental membrane. Therefore, for the pore formation we opted for a set of pore-lipid interaction parameters which on average generate a similar number of pores as in the experiment requiring smaller pore-lipid interaction parameters as one would expect in a natural membrane. Additionally, it was necessary to disallow direct pore-pore interactions, so as to avoid run-away aggregation of pores and infinite growth of the simulation matrix. The effect on the results should be negligible, as typically the pore density in the simulation was so low as to make pore-pore interactions insignificant. In the experiment collisions of pores are much less likely because of the much larger area, but may occur under stress when the membrane ruptures. The exact appearance of the pores vs. temperature profiles will, of course, depend on the choice of parameters. However, for a large range of interaction values, the permeability profile shows a strong peak at the phase transition temperature, and the position of this peak was found to always follow the peak of the heat capacity profile. Typical system sizes were 104 particles on a triangular lattice with periodic boundary conditions. The typical simulation was allowed to equilibrate for 2×10^4 to 10^5 Monte Carlo cycles (lattice sweeps) two times, first without pores involved, and then again with pore formation allowed. The actual sampling was then done over another 10⁵ to 5×10^{5} cycles. Most of the calculated curves presented were in addition averaged over several independent runs.

Our error analysis was performed using the blocking or bunching method described in Flyvbjerg and Petersen. Herror bars are not shown for heat capacity and pore numbers. The simulated c_p and pore numbers in Fig. 5, 6 and 8 represent the statistical mean and are not fitted. A detailed analysis can be found in the Masters thesis of A. Blicher. So

Some general consideration about the validity of lattice site simulations. Simulations based on an Ising model are successful for cooperative transitions with correlation length larger than individual lattice sites. Thus, they live on the fact that they don't require molecular detail. Therefore, one might argue that introducing a defect of the size of one lattice site is outside of the scope of such a simulation. This is partially true if one refers to molecular details about pore structure and geometry. It is not true if one attempts to make statements about the likelihood to obtain an area fluctuation of a particular magnitude. A pore in the matrix is most likely not just a hole in the lattice. On the other hand we know that the lipid ion channels display sizes of about one lipid diameter. There is some discussion in the literature about the geometry of such pores. Recent MD simulations suggest that the pore formation process is a formation of some aligned water molecules through the membrane followed by a lipid rearrangement where the lipid head groups point towards the aqueous water pore (cf. Fig. 1a and 1b). In our Monte Carlo simulation, the lipid rearrangement would be hidden in the interfacial energy of the hole (i.e., the hole-lipid interaction parameters). The second, important fact is that pore formation occurs by keeping the total area constant. Fluctuations in lipid area of the size of one lipid area, however, are completely within the scope of MC simulations as long as one does not attribute a molecular meaning to the pore geometry.

3 Results

In this paper we investigate pore formation in lipid membranes both experimentally with black lipid membranes (BLM) using the Montal–Mueller technique, ²⁶ and theoretically with Monte Carlo simulations. The goal is to demonstrate that pore formation is related to the heat capacity and the cooperative area fluctuations, respectively, and that anesthetics change the permeability in a coherent manner.

3.1 Black lipid membranes

In the following we show the conduction events through synthetic BLMs. We have chosen to work with lipid mixtures that have their transition events close to room temperature and that display broad heat capacity profiles. The latter makes it easier in experiments to adjust the temperature such that one is very close to the heat capacity maximum.

Fig. 2 shows the current traces for a DOPC:DPPC = 2:1 mixture at 19 °C (150 mM KCl, pH \approx 6.5) for various transmembrane voltages between -220 mV and +220 mV. The heat capacity profile for this lipid mixture can be found in Fig. 3A (bottom). It displays a c_p -maximum close to 19 °C. In agreement with earlier publications^{6,11} (e.g., we find quantized conduction events very similar to those reported for protein ion channels (Fig. 2, bottom) with voltage dependent amplitudes. The amplitudes display a linear dependence on the transmembrane voltage, V_m , between about -150 to +150 mV (Fig. 2, top). At voltages above $|V_m| = 150$ mV, the conductance is found to be higher than expected from the linear relation. The origin of this non-linear current–voltage relationship is not clear but may be related to the influence of voltage on the phase behavior of the lipid membranes. It has been shown by other authors that the position

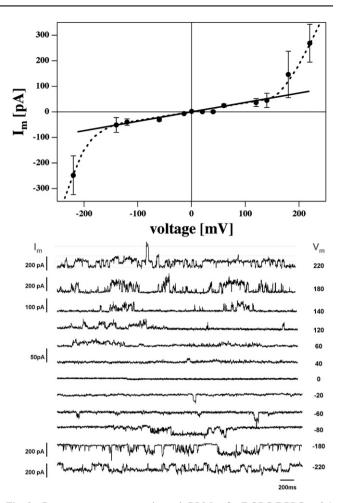


Fig. 2 Bottom: current traces through BLMs of a DOPC:DPPC = 2:1 mixture (150 mM KCl, pH ≈ 6.5 at T = 19 °C) for different voltages. Top: current–voltage relationship for the traces from the bottom panel. One finds a linear current–voltage relationship in the range from -150 mV to +150 mV. At higher voltages the traces become nonlinear. A total of 328 current traces with 30 s each where used for this graph. The current–voltage relationship was adapted from ref. 6.

of the melting transition can be influenced by voltage.^{36–38} Thus, we suspect that the nonlinear behavior at higher voltages is a consequence of the influence of voltage on the phase transition.

The effect of anesthetics. Fig. 3 shows the influence of octanol on the permeation of the above membranes measured at 19 °C and 210 mV. The top panels show heat capacity profiles (Fig. 3A), representative current traces (Fig. 3B), current histograms (Fig. 3C) and pore number probability distribution (Fig. 3D) in the presence of 15.9 mol% octanol in the membrane (calculated using an octanol partition coefficient in the membrane of 200).²⁷ The bottom panels show the respective data in the absence of octanol. One can see that the presence of octanol shifts the calorimetric profiles towards lower temperatures such that one is above the chain melting regime in the presence of octanol and in the transition regime in the absence of octanol. The current traces display smaller current amplitudes and a lower frequency of current events in the presence of octanol. Fig. 3C shows current amplitude histograms obtained

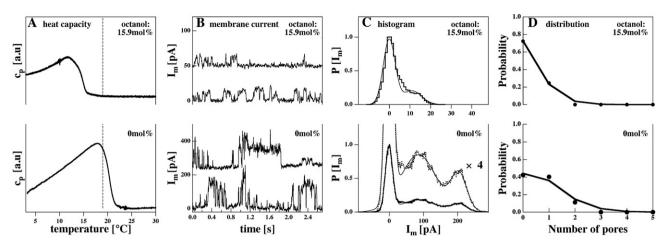


Fig. 3 Experiments on BLMs of a DOPC:DPPC = 2:1 mixture (150 mM KCl, pH ≈ 6.5) in the presence (top) and the absence (bottom) of 15.9 mol% octanol in the membrane. A: Heat capacity profiles. The dashed line indicates experimental temperature. Octanol shifts the calorimetric events towards lower temperatures. In the presence of octanol the experimental temperature above the melting events while it is at the c_p -maximum in the absence of octanol. B: Current traces obtained at T = 19 °C and 210 mV. Two representative traces are given for the two experimental conditions, respectively. Note the different amplitudes of the currents and the altered frequencies of current events. C: Current histograms from the curves in panel B showing one or two current events. D: Analysis of the frequencies of the number of pores. Symbols represent the areas of the peaks in panel C. The solid line represents a best fit to a Poisson distribution.

from the data shown in Fig. 3B. In the absence of octanol one finds at least two current steps corresponding to the two maxima of the histogram that are different from zero. In the presence of 15.9 mol% octanol one finds only one current step with a much lower amplitude. Fig. 3D shows the areas of the current peaks from Fig. 3C (obtained from fitting the peaks with Gaussian profiles). The solid lines are fits to a Poissonian distribution. Such distributions are expected for a low mean number of lipid ion channels.

Fig. 4 shows the corresponding experiments in the presence of ethanol. Ethanol is a very weak anesthetic with a membrane—water partition coefficient of about 0.48²⁸ (octanol has about

200). Three experiments recorded at $V_m = 120$ mV are shown, with 0 mol%, 9.5 mol% and 19.5 mol% ethanol in the fluid lipid membrane. Increasing amounts of ethanol shift the melting profile towards lower temperatures. The corresponding current traces show at least two current steps at 39 and 78 pA in the absence of ethanol (note that the voltage is lower than in the octanol experiment in Fig. 3), while there seems to be only one step for 19.5 mol% of around 45 pA. The current histograms suggests that the two current steps seen in the absence of ethanol merge into one in the presence of ethanol. This behavior is different from that of octanol where the overall conductance decreased upon the addition of anesthetics. We did not attempt

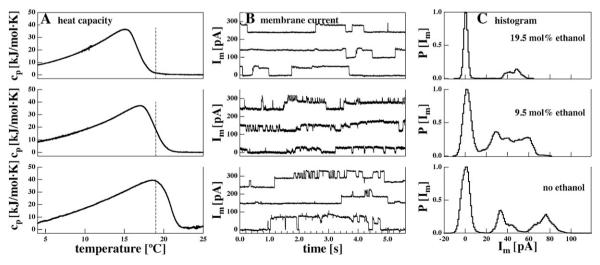


Fig. 4 Experiments on BLMs of a DOPC:DPPC = 2:1 mixture (150 mM KCl, pH \approx 6.5) in the presence of 19.5 mol% ethanol (top), 9.5 mol% (middle) and no ethanol in the fluid lipid membrane. A: Heat capacity traces. B: Current traces measured at 120 mV. The presence of ethanol seems to generate longer opening times. Three representative traces are given for each experimental condition. C: Histograms of the current events. The behavior is different as in the presence of octanol (Fig. 3). In the absence of ethanol one finds two clearly visible equally spaced current steps of about 39 pA that merges into one peak in the presence of ethanol.

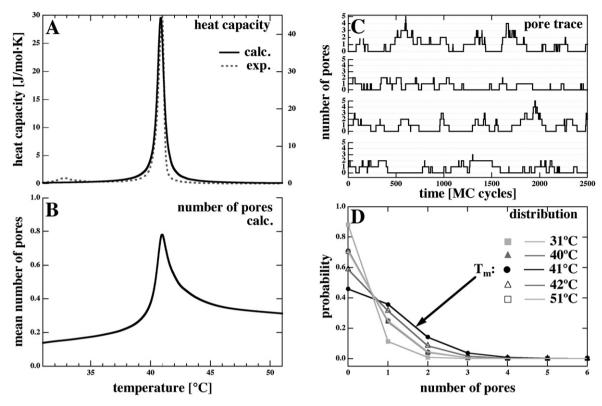


Fig. 5 Monte Carlo simulations of a DPPC membrane. A: calculated heat capacity profile compared to the experimental profile of DPPC large unilamellar vesicles. B: calculated temperature dependence of the membrane permeability (mean number of pores). C: four traces of the number of pores (corresponding to conductance) calculated at the melting temperature, T_m , at 41 °C. D: probability distribution of the number of pores. The symbols originate from the simulation, the solid lines are fits by a Poisson distribution. Simulation parameters are given in the text.

to make a Poisson analysis of the histograms of the ethanol containing membranes since the histograms in the presence of ethanol do not contain evenly spaced conduction levels in the presence of ethanol.

In the discussion section we show that the response of the lipid ion channels to both octanol and ethanol is very similar to that reported for protein ion channels.

3.2 Monte Carlo simulations

In the following we attempt to rationalize the experimental findings obtained from anesthetic-containing membranes by a statistical thermodynamics model that is described in detail in the Materials section. The main postulate in this simulation is to postulate a pore formation process that conserves the overall area of the total membrane. The conservation of area during the pore formation step implies that pores can only form when compressing the area of the lipids. Hence, the pore formation process is linked to the area fluctuations of the lipid matrix. The area fluctuations, however, are proportional to the lateral compressibility. Nagle and Scott³ argued that the likelihood of finding a pore is proportional to the compressibility. Now it is known that the excess heat capacity of lipid melting is proportional to the compressibility.9 In section 2.2 we have therefore argued that the likelihood of finding a pore must be intimately related to the heat capacity.6 Our Monte Carlo model makes use of the considerations of Nagle and Scott.3

Fig. 5 shows the results of a simulated DPPC membrane. The simulated heat capacity profiles are compared to an experimental profile of DPPC large unilamellar vesicles in Fig. 5A. The close agreement of the two curves is not accidental but a consequence of using the parameters ΔH , ΔS and the half width of the experimental heat capacity profile. Fig. 5B shows the calculated mean number of pores in the simulation box, which is proportional to the conductance of the membrane. They show that the mean number of simulated pores reaches a maximum close to the transition maximum at 41 °C. This behavior is exactly that found by Blicher et al.6 for dye permeation through large unilamellar vesicle membranes and confirms the concept of Nagle and Scott³ that makes use of the maximum of the lateral compressibility at T_m . Fig. 5C shows representative traces of the pore number computed at the melting temperature, T_m . These traces strongly resemble those from the BLM experiments (Fig. 2, 3 and 4). Fig. 5D shows the probability distribution of the number of pores for five different temperatures below, at and above the melting temperature. The solid lines in this panel represent Poisson distributions indicating that the calculated behavior matches the experimental one (Fig. 3D, bottom). The Poissonian statistics are expected when pores form independently of each other. The lipid-pore interaction parameters (ω_{pf} and ω_{pg}) are given in the Methods section and have been chosen such that the absolute number of pores is of similar order as in the BLM experiment. Since the simulated matrix has about 10 000 lipids while the black lipid membrane displays a much larger area of about 1.6 \times 10 10 lipids, the pore–lipid interaction parameters

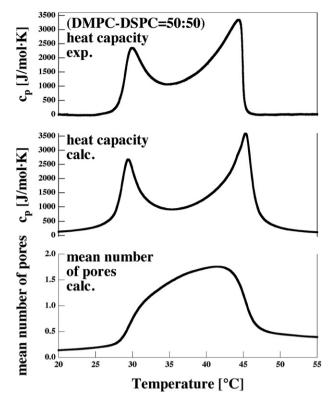


Fig. 6 DMPC–DSPC mixture. Top: experimental heat capacity profile. Middle: calculated heat capacity profile from MC simulation (shown is the statistical mean). Bottom: calculated mean number of pores as a function of temperature. One can recognize that the membrane permeability reaches a maximum in the chain melting regime. Simulation parameters are given in the text.

would have to be about 2–3 times higher if the simulation matrix had the same size as the BLM membrane. With such parameters, however, we would hardly ever see a conduction event in the simulation at a matrix size of 10 000 lipids. Therefore, we have chosen smaller parameters. Qualitatively, however, the behavior of the pore formation rate with a maximum at T_m is maintained when using the smaller model parameters given above.

The above concept also works in mixtures of different lipids with a broad melting peak. Fig. 6 shows the mean number of pores, *i.e.*, the permeation profile, for a DMPC–DSPC = 50:50 lipid mixture. The corresponding experimental and simulated heat capacity profile is shown in the top panels of Fig. 6. The maximum permeability (\equiv mean number of pores) is found in the chain melting regime between the two outer c_p maxima.

The effect of anesthetics. The famous Meyer–Overton rule³⁹ states that the effectiveness of anesthetics is exactly proportional to their solubility in lipid membranes and independent of their chemical nature. Further, anesthetics are known to lower the phase transition temperature of lipid membranes. In a recent paper we have shown that the effect of anesthetics on pure lipid membranes can be well described by the well-known freezing-point depression law $T_m = \frac{RT_m^2}{\Delta H} \cdot x_A$, where x_A is the molar fraction of anesthetics.¹⁸ The derivation of this formula is based on only two assumptions: the anesthetics molecule is ideally soluble in the liquid phase, and that it is insoluble in the solid phase. In

our simulation the anesthetics molecules are modeled as resembling lipids that occupy lattice sites. The perfect solubility of anesthetics in the fluid membrane phase can easily be modeled by assuming that the anesthetic–fluid interaction is zero ($\omega_{af}=0$), while the insolubility in the gel phase means that the anesthetic–gel parameter (ω_{ag}) is very large.

Fig. 7 shows Monte Carlo snapshots for DPPC membranes at seven different temperatures evenly spaced by 2.5 degrees. The three columns are calculated for 0 mol%, 5 mol% and 10 mol% anesthetics, respectively. Here, each anesthetic molecule occupies one lattice site. The melting temperature is again $T_m = 41$ °C. In this figure red and green dots represent gel and fluid lipids, while

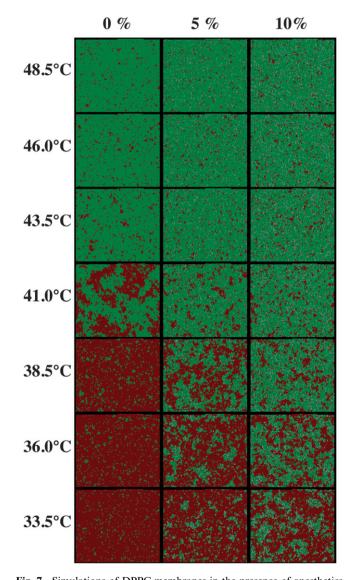


Fig. 7 Simulations of DPPC membranes in the presence of anesthetics as a function of temperature and anesthetics concentration. Left column: Monte Carlo snapshots in the absence of anesthetics. Center column: in the presence of 5 mol% anesthetics. Right column: in the presence of 10 mol% anesthetics. The rows indicate different temperatures between 33.5 °C and 48.5 °C. Gel lipids are given in red, fluid lipids in green, anesthetic molecules in white and pores in black. One can recognize that the presence of anesthetics shifts the melting temperature towards lower temperatures. Simulation parameters are given in the text.

white dots represent anesthetics molecules and black dots are pores. One can see how the increasing amount of anesthetic lowers the melting events as evident from the lower temperatures where one finds coexistence of red gel domains and green fluid domains. These simulations are further analyzed in Fig. 8. Fig. 8 (top) shows the heat capacity profiles for the simulations in

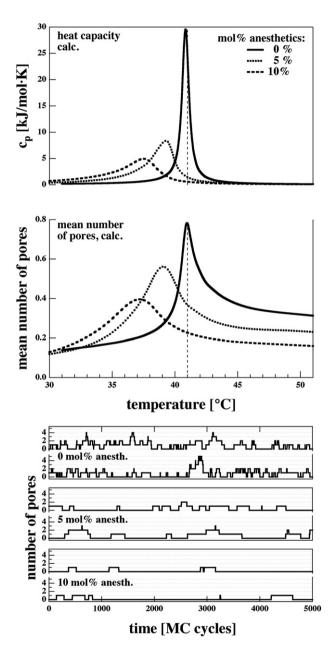


Fig. 8 Monte Carlo Simulation of DPPC membranes in the presence of 0 mol%, 5 mol% and 10 mol% of anesthetics in the membrane. Top: calculated heat capacity profiles show the shift of the melting events towards lower temperatures in agreement with the snapshots in Fig. 7. Middle: the calculated number of pores (corresponding to a permeability) follows the heat capacity profiles. The maximum permeability shifts towards lower temperatures upon addition of anesthetics. Bottom: calculated traces of the number of pores indicate that the number of pore opening events decreases in the presence of anesthetics. Two traces for each anesthetic concentration are given. Simulation parameters are given in the text.

Fig. 7. They are found to be very similar to those found experimentally in the case of DPPC with octanol18 and show the lowering of the melting temperature in the presence of anesthetics. Note, that the critical anesthetic dose for tadpoles corresponds to 2.6 mol% of anesthetics in the membrane, 18 which is somewhat lower but of same order as the concentrations used in this work. Fig. 8 (middle) shows the corresponding permeation profiles that show a similar shift towards lower temperatures in the presence of anesthetics. Fig. 8 (bottom) shows pore formation traces for three different anesthetics concentrations calculated at 41 °C. These traces correspond to the experimental situation in Fig. 3. The mean number of pores is reduced. Since Monte Carlo simulations are based on equilibrium thermodynamics the time scales of these traces do not correspond to a real time scale. This means that one cannot conclude from the simulation that the mean life time of the pore openings in the presence of anesthetics is longer as in the absence of anesthetics even though the traces seem to suggest that. This is an intrinsic feature of equilibrium statistical thermodynamics simulations that do not contain intrinsic time scales. However, the very procedure assumed for the pore generation process implies that the pore generation process must have a characteristic time scale that is related to the fluctuation life time of the lipid membrane. We have recently shown this to be the case for the pore formation process.7

4 Discussion

In this paper we investigate the lipid ion channel formation in pure lipid membranes in the absence and the presence of anesthetics. The goal was to demonstrate the origin of the pore formation from the cooperative area fluctuations. We studied pore formation both experimentally by the Montal-Mueller²⁶ black lipid membrane technique, and by Monte Carlo simulations. We show that pure lipid membranes in the absence of proteins display conduction events that resemble those seen in the presence of protein ion channels. For this reason we call them lipid ion channels. Such events have been shown before by a number of authors. 6,10-15,40-44 Although there are obviously many reports about such events it is surprising that such behavior is not better known. While the pores in lipid membranes are shown to exist in the absence of proteins, the conduction of ions through proteins cannot be measured in the absence of lipid membranes. Since the events are so similar to the protein ion channel recordings one requires an unambiguous criterion for how to distinguish the events originating from proteins and lipid membranes. To our knowledge, no such criterion exists. Possibly, site-specific mutations of proteins related to changes in channel activity can be considered as an indirect criterion, but this may not be satisfactory because this only demonstrates that the proteins are somewhat involved in the conduction process (cf. last paragraph of discussion).

The importance of phase transitions in lipid membranes for the permeability has been pointed out before.^{2–4,6,8,11} The cooperative fluctuation in area renders the pore formation more likely because the lateral compressibility is higher. This effect resembles the phenomenon of critical opalescence described by Einstein.²⁹ He calculated the work to move one molecule from one part of the volume to another in order to obtain the density fluctuations.

In a similar manner, the increased fluctuations in the membrane render it easier to move lipids such that pores can be created. In the present paper we used this concept to describe the pore formation process by a simple statistical thermodynamics model in which lipids are moved while simultaneously reducing the area of the surrounding lipids. One could also imagine fluctuations on larger length scales leading to the formation of pores of the size of one lipid, e.g. by switching 50 fluid and 50 gel lipids into 47 fluid and 53 gel lipids plus one pore. Thus, it may be sufficient to switch three fluid lipids somewhere in the matrix in order to create a hole. On the other hand, both Papahadjopoulos et al.² and Corvera et al.4 suggested that the pore formation process takes place at domain interfaces which are the regions of largest fluctuation. This is consistent with the finding that the conductance is maximum within the transition range. This is the view we have taken.

Since the change of the intensive thermodynamic variables (e.g., pressure, lateral pressure, chemical potentials of protons, of calcium or of anesthetic drugs) influences the melting behavior of membranes and thus the fluctuations in the membrane, it is straight forward to explore how changes in these variables influence permeability. For instance, it has been shown that voltage influences the phase transition temperature. 37,38 As evident in our measurements, the mean conductance through pores in lipid mixtures is linear in the range between about -150to +150 mV. For $|V_m| > 150$ mV the current-voltage relation becomes non-linear indicating that high voltage influences the state of the membrane and induces more conduction events. It has further been shown that changes in calcium concentration^{41,14} and pH changes^{12,13} influence the occurrence of lipid ion channels. Here, we have used the chemical potential of anesthetics in the membrane as a variable. Our present work is a continuation of a recent publication from our group⁶ that shows how anesthetics influence the permeation process by influencing the melting process. Here, we have demonstrated this for the case of octanol and ethanol, which both shift transitions to a similar degree when the membrane concentration of the drug is the same. Both drugs lead to a reduction of the overall conductance, but the modes of interaction seem different. A paper by Ly and Longo⁴⁵ described the interaction of short chain alcohols (methanol to butanol) on lipid membrane surface tension. The authors concluded that those alcohols are surface active. Thus, the mode of ethanol interaction with membranes may contain other contributions as compared to octanol. The difference of the ethanol influence is visible in the fact that the overall conductance does not change in the same manner as in the presence of octanol. Instead, two conduction steps merge into one. Thus, we focused in our simulations on octanol which displays a very straight-forward mode of action, namely freezing point depression.

Our simulations describe the main feature found in experiments: the relation of the permeability and the heat capacity leading to maximum permeability at the heat capacity maximum. This confirms the main assumption of our study, which is that the cooperative area fluctuations are responsible for the occurrence of the lipid ion channels. It further describes quantitatively the effect of anesthetics on the heat capacity of the membrane. Anesthetics following the Meyer–Overton correlation also generate freezing-point depression (cf. ref. 18). The simulation

accurately describes this and therefore also describes the influence of anesthetics on the permeation. However, our pore formation model made the simplifying assumption that a pore has the size of one lipid. This is a reasonable approximation since we found in a recent paper that the conductance through the membrane is well described by assuming an aqueous pore of a diameter of 0.7 nm, which is of same order than that of one lipid site. The measurements in the presence of anesthetics show. however, that the pore size itself can vary and that it is different in the presence and absence of octanol and ethanol. This effect is not contained in our simulations. The pore diameter scales with the square root of the current. In the case of octanol the conductance in the presence of 15.9% octanol is about 7 times smaller than in the absence of octanol. This suggests a pore diameter that is about 2.6 times smaller as the pores in pure membranes. However, such calculations should be taken with care as they assume that the pore interior is filled with free water with a viscosity unaffected by the pore interface, which we consider to be unlikely, both for lipid ion channels as for protein channels. If water ordering at the pore interface is taken into account, the real pore radius may be larger, and the difference in the channel conductances may not be proportional to the pore area since interfacial effects will be larger for smaller pores.

The present Monte Carlo simulation is a coarse grain statistical thermodynamics model. Such models have successfully been used to describe cooperative phenomena in membranes. 19-24 Typically, the number of states for each lipid is reduced to two (Ising model) or 10 (Pink model). Naturally, this is only a very approximate description on the molecular scale. The reason why the lattice models work so nicely is that they are designed to describe cooperative behavior of many molecules—in our case the formation of domains and the cooperative fluctuations. Microscopic details of single lipids and pores described here by lattice sites are not expected to be accurate—and also to be of small relevance. As mentioned above, our simulation is designed such that a pore has always the size of one lattice site. The experiment on octanol shows that the pores in the presence of anesthetics are smaller as in the absence (currents at identical voltage are $\approx 90 \text{ pA}$ in the absence of octanol and only about \approx 15 pA in the presence of 15.9 mol% octanol). Qualitatively, however, the simulation describes the likelihood of pore formation and thereby permeation quite well. In a recent paper we have shown that the permeation of fluorescence dyes through lipid vesicles is within experimental accuracy proportional to the heat capacity. This paper also shows the quantized conduction events in BLMs that could be abolished by addition of octanol. Our simulation reproduces this close relation between heat capacity and conductance—and is consistent with the quantized nature of the conduction process.

It is likely that many other variable changes would influence the occurrence of lipid ion channels, *e.g.*, proteins that associate or insert into membranes but are not pores in themselves. It is known that the band 3 protein of erythrocytes increases the melting temperature of model membranes. 46 Cytochrome b₅, on the other hand lowers transitions. 47 Cytochrome c binds peripherally and increases the transition temperature. 22 Depending on whether the experimental temperature is found below or above the melting temperature, one would expect that the respective proteins can either increase or lower the likelihood

of lipid ion channel formation even if the proteins are not known to form ion channels.

The mechanism of anesthesia has been under much debate and is unexplained until today. Until the mid 1970's lipid models were most popular because they can be easily related to the famous Meyer-Overton rule39 that states that the effectiveness of anesthetics and their partition coefficient are linearly related, independent of the chemical nature of the anesthetic molecules. Ever since the development of the patch clamp methods and the study of protein ion channel events, a lot of attention has been on the binding of anesthetics to receptors. This is due to the finding that ion channel events that have been attributed to proteins are influenced by anesthetics. Unfortunately, such interaction with proteins is not generally in agreement with the Meyer-Overton correlation. The fact that the noble gas xenon acts as an anesthetic is inconsistent with specific binding to a protein. We have shown here and in ref. 6 that the finding of quantized events does not prove the action of a protein. Further, the influence of anesthetics on the conduction events does not prove a specific binding of the anesthetic drug to a receptor. We have shown here that octanol and ethanol influence the channel events even in the complete absence of proteins.

In this context it is tempting to compare the effect of alkanols on protein ion channels with our findings. The influence of octanol on current fluctuations of lipid ion channels can be compared with those of neuronal nicotinic acetylcholine receptor (AChR, see, e.g., ref. 48), which has been used for many years as a model to study the influence of alcohols. 49-55 It has been shown that neuronal activity of nicotinic acetylcholine receptor (AChR) induced channels under the influence of octanol is markedly inhibited, 48 which was observed as a decrease in current fluctuation amplitude from those channels. Ethanol, on the other hand was found not to be neutral for many protein ligand-gated ion channels such as N-methyl-D-aspartate (NMDA), serotonin (5-HT(3)), glycine and GABA receptors. 56,57 The inhibitory action of ethanol was also observed in voltage-gated Ca2+ channels⁵⁸ and very recently also in potassium channels, which were seen as very important targets of ethanol.⁵⁹ Similar findings were reported for experiments with AChR channels subjected to 100 mM of ethanol.60 The above reports demonstrate that the effect of octanol and ethanol on membrane protein receptors is quite similar to that found by us for the effect of these alcohols on lipid ion channels. In this context it is also interesting to note that protein ion channels have been shown to respond to the phase behavior of lipids⁶¹ or to different lipid environments.⁶²

Summarizing, one has to conclude that many channel events that have been attributed to particular specific proteins may well originate from membrane fluctuations. The voltage-clamp experiment in itself can not distinguish proteins from lipid events. If the observed currents in fact originated from the lipid membrane, the dependence of the open-probabilities of channels on the various intensive variables as pressure (mechano-sensitive channels), temperature (temperature sensors) and drugs and ions (channel gating) would immediately find an explanation in the well-understood thermodynamics of cooperative membrane events.

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